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ISOM-671: Managing Big Data

October 2nd, 2023

Group Assignment 1: DW and CAP Theorem

Part 1: Advanced SQL (movie recommendation system for Sakila):

1. Create a table of cosine similarity (user-A, user-B, similarity_score). You can estimate this similarity as: number of same movies rented by any two customers (A and B) divided by sqrt(total movie rented by user-A * movie rented by user-B)

SQL Query:

```
with rented_films as ( select distinct c.customer_id, i.film_id

from customer c left join rental r on c.customer_id = r.customer_id left join

inventory i on r.inventory_id = i.inventory_id),

customer_films_ordered as ( select distinct customer_id, film_id,

count(film_id) over (partition by customer_id) cust_films_ordered from rented_films),

movie_combinations as ( select distinct a.customer_id a_cust_id,

b.customer_id b_cust_id, b.film_id, a.cust_films_ordered a_cust_films,

b.cust_films_ordered b_cust_films from customer_films_ordered a

inner join customer_films_ordered b on a.film_id = b.film_id

and a.customer_id < b.customer_id)

select a_cust_id, b_cust_id, count(distinct film_id)/sqrt(a_cust_films*b_cust_films) as

cosine_similar from movie_combinations

group by a_cust_films, b_cust_films, a_cust_id, b_cust_id order by cosine_similar desc;
```

First 10 rows.

| | □ a_cust_id ÷ | □ b_cust_id ÷ | □ cosine_similar ÷ |
|----|---------------|---------------|---------------------|
| 1 | 24 | 111 | 0.28 |
| 2 | 53 | 185 | 0.26648544566940835 |
| 3 | 217 | 433 | 0.2502172968684897 |
| 4 | 371 | 542 | 0.24618298195866545 |
| 5 | 19 | 491 | 0.24535824603285922 |
| 6 | 201 | 317 | 0.23533936216582083 |
| 7 | 150 | 292 | 0.2309401076758503 |
| 8 | 18 | 109 | 0.21398024625545645 |
| 9 | 237 | 376 | 0.21134098610290408 |
| 10 | 350 | 495 | 0.20851441405707477 |

2. For any selected customer (user-A), and based on similarity score from previous step, recommend a movie not watched by the user-A.

SQL Query

```
with users as (select a cust id, b cust id, cosine similar from cosine similarity
    union all select b cust id, a cust id, cosine similar from cosine similarity),
highest similarity as (select distinct a cust id,
    max(cosine similar) over (partition by a cust id) max similar from users),
    pair with max score as (select a cust id, b cust id, cosine similar from (select
hs.a cust id, u.b cust id, u.cosine similar,
         row number() over (partition by hs.a cust id) as rn from highest similarity hs left join
users u on hs.a cust id = u.a cust id and hs.max similar = u.cosine similar) as e where rn = 1
movies rented as (select distinct c.customer id, i.film id from customer c left join rental r
    on c.customer id = r.customer id left join inventory i on r.inventory id=i.inventory id),
movie recommendation as ( select *,
    (select distinct film id from movies rented where customer id = pwms.b cust id and
film id not in (select film id from movies rented where customer id = pwms.a cust id) limit 1)
recommended movie from pair with max score pwms)
select mr.a cust id, c.first name, f.film id recommended film id,
title recommended film title from movie recommendation mr left join film f on
mr.recommended movie = f.film id left join customer c on a cust id = c.customer id;
```

| □ a_cust_id ÷ | ☐ first_name ÷ | ☐ recommended_film_id ÷ | ☐ recommended_film_title |
|---------------|----------------|-------------------------|--------------------------|
| 1 | MARY | 483 | JERICHO MULAN |
| 2 | PATRICIA | 745 | ROSES TREASURE |
| 3 | LINDA | 39 | ARMAGEDDON LOST |
| 4 | BARBARA | 887 | THIEF PELICAN |
| 5 | ELIZABETH | 557 | MANCHURIAN CURTAIN |
| 6 | JENNIFER | 454 | IMPACT ALADDIN |
| 7 | MARIA | 415 | HIGH ENCINO |
| 8 | SUSAN | 197 | CRUSADE HONEY |
| 9 | MARGARET | 730 | RIDGEMONT SUBMARINE |
| 10 | DOROTHY | 295 | EXPENDABLE STALLION |

Part 2: Data Warehouse - Star Schema (for real-estate investment planning):

yelp business hours

LOAD DATA LOCAL

INFILE

'/Users/FrancisJingo1/Desktop/Emory/courses/Fall_2023/ISOM_671_Managing_Big_Data/Group_assig ment/archive/yelp business hours.csv'

INTO TABLE yelp business hours

FIELDS TERMINATED BY ',' OPTIONALLY ENCLOSED BY ""

LINES TERMINATED BY '\n' IGNORE 1 ROWS

(business_id, monday, tuesday, wednesday, thursday, friday, saturday, sunday);

| | 📭 business_hour_id 🗧 📭 business_id | ÷ □ monday | ÷ □ tuesday | ÷ Ⅲ wednesday | ÷ □ thursday | ≎ [∐ friday | ≑ 🏻 saturday | ≎ 🛘 sunday 🗢 |
|---|------------------------------------|------------|-------------|---------------|--------------|-------------|--------------|--------------|
| 1 | 1 FYWN1wneV18bWNgQjJ26Ng | 7:30-17:0 | 7:30-17:0 | 7:30-17:0 | 7:30-17:0 | 7:30-17:0 | None | None |
| 2 | 2 He-G7vWjzVUysIKrfNbPUQ | 9:0-20:0 | 9:0-20:0 | 9:0-20:0 | 9:0-20:0 | 9:0-16:0 | 8:0-16:0 | None |
| 3 | 3 KQPW8lFf1y5BT2MxiSZ3QA | None | None | None | None | None | None | None |
| 4 | 4 8DShNS-LuFqpEWIp0HxijA | 10:0-21:0 | 10:0-21:0 | 10:0-21:0 | 10:0-21:0 | 10:0-21:0 | 10:0-21:0 | 11:0-19:0 |
| 5 | 5 PfOCPjBrlQAnzNXj9h_w | 11:0-1:0 | 11:0-1:0 | 11:0-1:0 | 11:0-1:0 | 11:0-1:0 | 11:0-2:0 | 11:0-0:0 |

yelp business attributes

LOAD DATA LOCAL

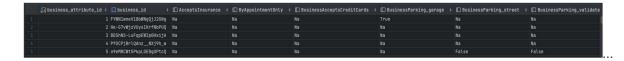
INFILE

'/Users/FrancisJingo1/Desktop/Emory/courses/Fall_2023/ISOM_671_Managing_Big_Data/Group_assig ment/archive/yelp business attributes.csv'

INTO TABLE yelp business attributes

FIELDS TERMINATED BY ',' OPTIONALLY ENCLOSED BY ""

LINES TERMINATED BY '\n' IGNORE 1 ROWS;



LOAD DATA LOCAL INFILE

'/Users/FrancisJingo1/Desktop/Emory/courses/Fall_2023/ISOM_671_Managing_Big_Data/Group_assig ment/archive/yelp_review.csv'

INTO TABLE yelp review

FIELDS TERMINATED BY ',' OPTIONALLY ENCLOSED BY "" ESCAPED BY '\b' LINES TERMINATED BY '\r\n' IGNORE 1 ROWS;

| | ∏review_id ÷ | □user_id : | D business_id : | □stars : □date : | □ text : | □ useful : | □ funny : □ cool : |
|---|---------------------|------------------------|------------------------|------------------|---|------------|--------------------|
| 1 | Bw8LtQgezPiN9xJWaQ | lMkqkljZsQ1pm0Zb_bbP8A | jCNBZnkIFv_0omLVTgNR6Q | 5 2017-10-11 | Don't know how I missed this place after so many years here, but really fabulous | | |
| 2 | 05rSAAHBiM7XAbXsW-A | KPPOpDFY05HB0V0dFgSDWw | 9Q1ZtzTPFWG4fJiFSko5Xg | 4 2011-89-21 | I visited Cantina Laredo for a Sunday Brunch and I have yet to get it off my min | | |
| 3 | 0XFGhj0U1H8Y3cVYjMA | OsFWc7PMDDACG9MMit7kGQ | oWboXKe_xk6Vcr2gBEuxuw | 2 2014-05-11 | Be aware. There is an extremely limited menu. 443 pastas-2 sauces- marinara or sa | | |
| 4 | 3SR6DPz0F6gLBxxjuVw | vHF4LqmMRkhrLD5FE-8HXA | TgEKtJGC-cN9rrCKgSDx8g | 5 2017-09-23 | This place is amazing. aThe strawberry cheesecake with cream is my favorite !aTh | | |
| 5 | 4_AFJm_f0E-HTgPDxjw | 6mn-M3f75hdynz245p-fBA | q7MorRPzU_J-iekeDKUKgw | 5 2011-02-02 | i really liked this place⊲tequila bar!! how awesome!!⊲i like how little this p | | |

yelp checkin

LOAD DATA LOCAL

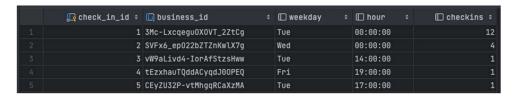
INFILE

'/Users/FrancisJingo1/Desktop/Emory/courses/Fall_2023/ISOM_671_Managing_Big_Data/Group_assig ment/archive/yelp_checkin.csv'

INTO TABLE yelp checkin

FIELDS TERMINATED BY ',' OPTIONALLY ENCLOSED BY "" ESCAPED BY 'b'

LINES TERMINATED BY '\n' IGNORE 1 ROWS;



yelp tip

LOAD DATA LOCAL

INFILE

'/Users/FrancisJingo1/Desktop/Emory/courses/Fall_2023/ISOM_671_Managing_Big_Data/Group_assigment/archive/yelp_tip.csv'

INTO TABLE yelp tip

FIELDS TERMINATED BY ',' ENCLOSED BY "" ESCAPED BY '\b'

LINES TERMINATED BY '\r\n' IGNORE 1 ROWS;



yelp business

INFILE

'/Users/FrancisJingo1/Desktop/Emory/courses/Fall_2023/ISOM_671_Managing_Big_Data/Group_assig ment/archive/yelp_business.csv'

INTO TABLE yelp business

FIELDS TERMINATED BY ',' OPTIONALLY ENCLOSED BY ''' ESCAPED BY '\b' LINES TERMINATED BY '\r\n' IGNORE 1 ROWS;

| ∰ business_id ÷ | □ name ÷ | □ neighborhood ÷ | □ address = | | ∷ 🗓 state : | postal_code : | □ latitude : | □ longitude : | □ stars : | ☐ review_count : | □ is_open | : □ categories |
|-----------------------|-----------------------------|------------------|--------------------------|-----------|-------------|---------------|---------------|----------------|-----------|------------------|-----------|------------------------------------|
| 11u67MLxW6FIv2fC6PiQQ | "SpinalWorks Chiropractic" | | "15648 N 7th St, Ste A3" | Phoenix | | 85822 | 33.628854 | -112.8659754 | | | | 1 Physical Therapy; Chiropractors; |
| 23I-DDkqM9XjLH1cJl3VA | "Montallegro Barber Shop" | Villeray-Saint-M | "7244 Rue Hutchison" | Montreal | | H3N 1Z1 | 45.5298637 | -73.6237259 | | | | 1 Hair Salons;Barbers;Beauty & Sp |
| 33q0wWFBUE8md0ToI7YrQ | "Custom Kings" | Southeast | | Las Vegas | | 88981 | 36.0556569492 | -115.169422894 | | | | 1 Screen Printing/T-Shirt Printir |
| 447_7H-yK3HCh05vyut_Q | "Instant Muffler and Autore | | "1295 Weston Road" | | | M6M 4R2 | 43.6892366 | -79.4952863 | | | | 1 Auto Repair; Automotive |
| 56jYJ6Hm-Qq8XQEGDr0GQ | "Winfield Gene DO" | | | Тепре | | | 33.4055938 | -111.9394369 | | | | 1 Doctors;Health & Medical |

yelp user

LOAD DATA LOCAL

INFILE

'/Users/FrancisJingo1/Desktop/Emory/courses/Fall_2023/ISOM_671_Managing_Big_Data/Group_assig ment/archive/yelp_user.csv'

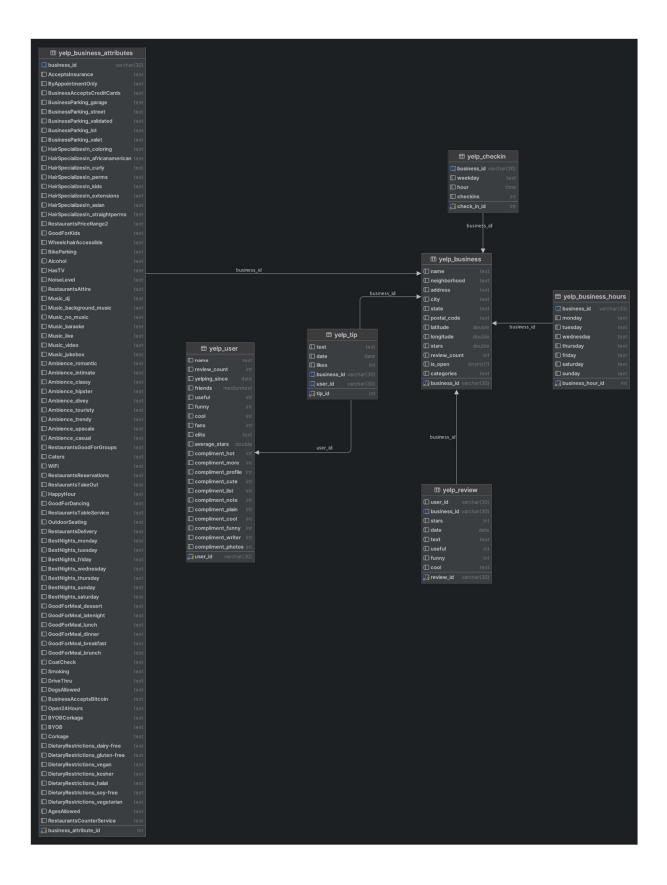
INTO TABLE yelp_user

FIELDS TERMINATED BY ',' ENCLOSED BY "" ESCAPED BY '\b'

LINES TERMINATED BY '\r\n' IGNORE 1 ROWS;

| ∰ user_id | : □ name : | ☐ review_count : ☐ yelping_since : | □ friends | □ useful ÷ | □ funny ÷ | □ cool ; | □ fans : | □ elite |
|----------------------|------------|------------------------------------|---|------------|-----------|----------|----------|---------|
| 1DPmKJsBF2X6ZKgAeGqg | Charlotte | 3 2014-04-07 | None | | | | | 3 None |
| 2fEWlObjtPaZ-pK0eq9g | Nerissa | 7 2816-85-89 | Ngwaot7XkD4g75hHBY3wnQ | | | | | 9 None |
| 3I9ZYdYGkZ6dMYxwJEIQ | | 168 2811-88-17 | JWrq6BEnAFoRZ4zDmlB8Yg, h0E5zGo6HVQGn0fYiRyMPg, 9e-TKXZf2nf0M9dwxkL_KQ, V8-d80YZs_uo1BR8W0f | | | | | 7 None |
| 4MTsBloH4jvybJ5DrTYw | | 10 2011-03-25 | KN6I2UWhB-N2nXtaeLgNsg, R3ZzMoqN3FibAu810078rg, dxaz3o-XoRyvdUhdDqr8CA, MlsidvbvzPNmxTo7n83 | | | | | None |
| 5QCazm0YrHLd3uNUPYMA | | | None | | | | | 3 None |

PART 2.1: ER Model



1. Create fact and dim tables (in star-schema) using SQL queries (submit all SQL queries and resulting ER models). For the fact table, consider your role as a commercial real-estate investor who is interested in foot traffic and quality of business in various zip codes. Your goal is to identify what areas to invest in real-estate - i.e., you like to query the ratings and check-ins by date for all businesses in a zip code. (e.g., how many people checked in to businesses in Atlanta during summer, and what ratings they received?)

SQL QUERIES

```
create table dim_location as
select row_number() over () location_id, neighborhood, address,
city, state, latitude, longitude from
(select distinct neighborhood, address, city, state,
latitude, longitude from yelp_business) as e;
```

```
create table dim_business as
select yb.business_id, yb.name, yb.is_open,
categories, monday, tuesday, wednesday, thursday, friday, saturday, sunday
from yelp_business yb left join yelp_business_hours ybh
on yb.business_id = ybh.business_id;
```

```
create table dim_user as select * from yelp_user;
```

```
create table dim_tip select * from yelp_tip;
```

```
create table dim review daily as select * from yelp review;
```

create table dim checkin select * from yelp checkin;

```
create table business_fact as

select row_number() over () business_fact_id, yb.business_id,

yb.stars, yb.review_count, drd.date, dl.location_id, dc.check_in_id,

drd.review_id, du.user_id, yb.postal_code, dt.tip_id

from yelp_business yb left join dim_business db

on yb.business_id = db.business_id left join dim_location dl on

yb.longitude = dl.longitude and yb.latitude = dl.latitude and

yb.neighborhood = dl.neighborhood and yb.city = dl.city and yb.state = dl.state

and yb.address = dl.address left join

dim_checkin dc on db.business_id = dc.business_id left join dim_review_daily drd

on yb.business_id = drd.business_id left join dim_tip dt on db.business_id = dt.business_id

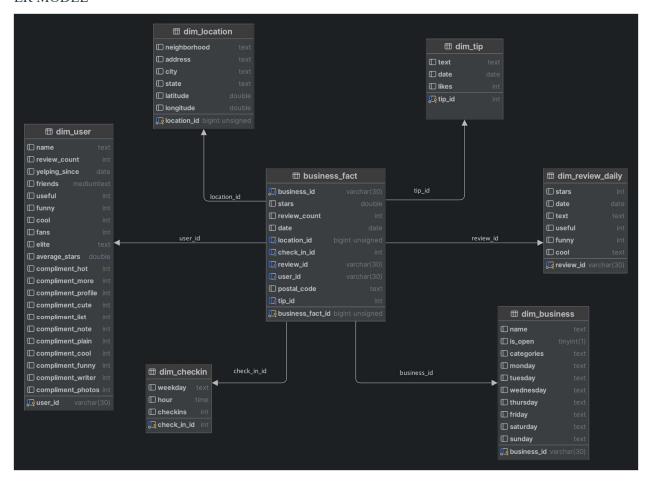
left join dim_user du on drd.user_id = du.user_id;
```

```
# removing the business_id column from other tables
alter table dim_checkin drop business_id;
alter table dim_review_daily drop business_id;
alter table dim_review_daily drop user_id;
alter table dim_tip drop business_id;
alter table dim_tip drop user_id;
```

```
# adding primary keys
alter table business_fact add primary key (business_fact_id);
alter table dim_business add primary key (business_id);
alter table dim_checkin add primary key (check_in_id);
alter table dim_review_daily add primary key (review_id);
alter table dim_tip add primary key (tip_id);
alter table dim_location add primary key (location_id);
alter table dim_user add primary key (user_id);
```

```
# adding foreign keys
alter table business_fact add foreign key (business_id) references dim_business(business_id);
alter table business_fact add foreign key (check_in_id) references dim_checkin(check_in_id);
alter table business_fact add foreign key (review_id) references dim_review_daily(review_id);
alter table business_fact add foreign key (tip_id) references dim_tip(tip_id);
alter table business_fact add foreign key (location_id) references dim_location(location_id);
alter table business_fact add foreign key (location_id) references dim_location(location_id);
alter table business_fact add foreign key (location_id) references dim_location(location_id);
alter table business_fact add foreign key (location_id) references dim_location(location_id);
alter table business_fact add foreign key (user id) references dim_user(user id);
```

ER MODEL



Part 3: Scaling Yelp

Scaling Yelp globally would imply extending their current product offerings to a wider market. Since consumer trends and reviews are tied exclusively to location, the most efficient structure to scale Yelp would be to group information by geographical region. Document based systems could accurately represent differences in location and what consumers consider essential features of a business. In document-based NoSQL databases, data is stored in the form of documents which allows for ease of access and retrieval using ids. This unstructured approach to database systems allows for flexibility between dimensions of geography for Yelp's global scale. Transitioning to NoSQL database systems is an intuitive choice when considering that reviews are composed primarily of unstructured data. The heading and body of a review are unstructured data because they can vary in length, content, format, and style with no predefined model.

CAP Theorem suggests that a distributed system can only sustain two of the following characteristics: consistency, availability, and partition tolerance. In the context of upscaling Yelp's business model globally, we believe that Yelp should prioritize availability and partition tolerance for their databases. Availability refers to the fact that Yelp's servers must be able to fulfill any request or output a message which claims the message can't be completed. For Yelp's segment of customers who use the app to guide their decision-making process, being able to access and retrieve information at any point is invaluable. It is practically impossible for a globally scaled company to operate without partition tolerance, as segments located in different regions should have the capability to operate independently of each other. While this design choice sacrifices consistency, we believe that the business value of on-demand access to information outweighs the cost of lack of consistency across nodes.